WEST

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**Search Results -**

| Term | Documents |
|---|---|
| "6363333" | 2 |
| 6363333S | 0 |
| "6363333".UREF..USPT,PGPB,JPAB,EPAB,DWPI,TDBD. | 1 |
| (6363333.UREF.).USPT,PGPB,JPAB,EPAB,DWPI,TDBD. | 1 |

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## Search History

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|---|---|---|---|
| *DB=USPT,PGPB,JPAB,EPAB,DWPI,TDBD; PLUR=YES; OP=ADJ* | | | |
| L6 | 6363333.uref. | 1 | L6 |
| L5 | L4 and (time adj5 interval$1) | 0 | L5 |
| L4 | (time and stock and market$).ti. | 21 | L4 |
| L3 | (interval$ and track$ and financi$).ti. | 0 | L3 |
| L2 | L1 and time | 1 | L2 |
| L1 | (interval$ and financial and data$).ti. | 3 | L1 |

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| *DB=USPT,PGPB,JPAB,EPAB,DWPI,TDBD; PLUR=YES; OP=ADJ* | | | |
| L26 | 5671363.uref. | 17 | L26 |
| L25 | 5671363.pn. | 2 | L25 |
| L24 | (stock$1 and trad$ and database$).ti. | 4 | L24 |
| L23 | (stock$1 and trad$ and sql).ti. | 0 | L23 |
| L22 | 5946666.pn. | 2 | L22 |
| L21 | (stock$1 and market$ and analysis).ti. | 17 | L21 |
| L20 | 6199077.pn. | 2 | L20 |
| L19 | L17 and (dynamic near5 updat$) | 4 | L19 |
| L18 | L17 and (stock nar5 table$1) | 0 | L18 |
| L17 | L16 and (stock near5 track$) | 24 | L17 |
| L16 | L15 and (stock near5 symbol$1) | 201 | L16 |
| L15 | stock near5 market$ | 3463 | L15 |
| L14 | (stocks and securities and track$).ti. | 1 | L14 |
| L13 | (stocks and securities and trend$).ti. | 0 | L13 |
| L12 | (stocks and securities and real and time).ti. | 3 | L12 |
| L11 | (time and serie$ and financial and data$).ti. | 2 | L11 |
| L10 | L9 and (interval near5 database$) | 3 | L10 |
| L9 | L8 and (stock near5 database$) | 57 | L9 |
| L8 | (financial near5 database$) | 1688 | L8 |
| L7 | (stock and market$ and real and time).ti. | 6 | L7 |
| L6 | L5 and (time near5 interval$1) | 1 | L6 |
| L5 | raw financial data | 13 | L5 |
| L4 | L3 and (query$ or search$) | 6 | L4 |
| L3 | l1 and (time adj5 interval$1) | 22 | L3 |
| L2 | time interval data tables | 1 | L2 |
| L1 | raw data tables | 79 | L1 |

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| L10 | L9 and (financial near5 markets) | 0 | L10 |
| L9 | ((plurality near5 intervals) same (adj$ near5 data)) | 67 | L9 |
| L8 | L7 and ((plurality near5 intervals) same (adj$ near5 data)) | 0 | L8 |
| L7 | (financial and market$).ti. | 183 | L7 |
| L6 | 5161103.uref. | 15 | L6 |
| L5 | L4 and (time near5 interval$1) | 7 | L5 |
| L4 | (time near5 vary$) same (stock near5 data$) | 12 | L4 |
| L3 | L1 and (real near5 time) | 3 | L3 |
| L2 | L1 and ((time near5 vary$) same (volume near5 data)) | 0 | L2 |
| L1 | (stock price$) same (clos$ near5 open$) | 18 | L1 |

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CiteSeer Find: time series market trends | Documents | Citations

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Clustering Time Series with Hidden Markov Models and.. - Oates, Firoiu, Cohen (1999) (Correct) (1 citation)
Clustering **Time Series** with Hidden Markov Models and Dynamic **Time**
www-eksl.cs.umass.edu/papers/oates-ijcai99SL.ps

Parallel and Distributed Search for Structure in.. - Oates, Schmill, Cohen (1996) (Correct) (2 citations)
Distributed Search for Structure in Multivariate **Time Series** Tim Oates, Matthew D. Schmill and Paul R.
Search for Structure in Multivariate **Time Series** Tim Oates, Matthew D. Schmill and Paul R. Cohen
as those describing the ebb and flow of the stock **market** or the health of a patient in an intensive care
www-eksl.cs.umass.edu/papers/Oates96a.ps

Investigation of Periodic Time Series using Neural Networks.. - Gregory Noone (1995) (Correct) (1 citation)
Investigation of Periodic **Time Series** using Neural Networks and Adaptive Error
wwwcrasys.anu.edu.au/PTP/Projects/pulseTrain/Projects/pulseTrain/Projects/pulseTrain/Papers/../Papers/NH95b.ps

Transionospheric Signal Detection with Chirped Wavelets - Doser, Dunham (Correct)
utilized to detect dispersed signals in the joint **time**/scale domain. Specifically, pulses that become
discrete wavelet transform, applied to actual **time series** recorded by the US Department of Energy's
www.utdallas.edu/~doser/as97paper.ps

A Neighborhood Map of Competing One Step Predictors for.. - Fancourt, Principe (Correct)
for Piecewise Segmentation and Identification of **Time Series** Craig L. Fancourt and Jose C. Principe
Piecewise Segmentation and Identification of **Time Series** Craig L. Fancourt and Jose C. Principe
www.cnel.ufl.edu/bib/papers/fancourt96icnn.ps.gz

Markov Switching Time Series Models with Application to a.. - Lu, Berliner (1999) (Correct) (2 citations)
Markov Switching **Time Series** Models with Application to a Daily Runoff
Markov Switching **Time Series** Models with Application to a Daily Runoff **Series**
www.cgd.ucar.edu/stats/papers/lu_berliner.ps.Z

Radar Pulse Train Parameter Estimation and Tracking using.. - Greg Noone (1995) (Correct)
network is used based on a simple state space **time series** formulation of the radar problem. The
network is used based on a simple state space **time series** formulation of the radar problem. The network
chaotic analytic functions as well as some "stock-**market**" type problems [1, 2]**Time series** network
wwwcrasys.anu.edu.au/PTP/Projects/pulseTrain/Projects/pulseTrain/Papers/../Papers/Noo95.ps.gz

A Componentized Architecture for Dynamic Electronic Markets - Reich, Ben-Shaul (1998) (Correct) (2 citations)
bids and asks are collected for a predetermined **time** interval and are matched at the end of the
A Componentized Architecture for Dynamic Electronic **Markets** Benny Reich Israel Ben-Shaul Department of
www.dsg.technion.ac.il/gem/papers/gem-sigmodrec.ps.gz

TREND: A System for Generating Intelligent Descriptions of.. - Sarah Boyd (1998) (Correct) (1 citation)
A System for Generating Intelligent Descriptions of **Time-Series** Data Sarah Boyd Microsoft Research
for Generating Intelligent Descriptions of **Time-Series** Data Sarah Boyd Microsoft Research Institute
**TREND**: A System for Generating Intelligent Descriptions
www.mri.mq.edu.au/~sarahb/icips.ps

Change of structure in financial time series, long range.. - Mikosch, Starica (1999) (Correct) (2 citations)
Change of structure in financial **time series**, long range dependence and the GARCH model
www.cs.rug.nl/~eke/iwi/preprints/99-5-06.ps.gz

Modelling and robustness issues in Bayesian time series analysis - West (1995) (Correct)
Modelling and robustness issues in Bayesian **time series** analysis Mike West ISDS, Duke University,
Modelling and robustness issues in Bayesian **time series** analysis Mike West ISDS, Duke University,
ftp.stat.duke.edu/pub/WorkingPapers/95-12.ps

Flexible Seasonal Long Memory and Economic Time Series - Marius Ooms (1995) (Correct) (3 citations)
Flexible seasonal long memory and economic **time series** Marius Ooms October 12, 1995 Econometric
www.eur.nl/few/eb/papers/../pub/oomsart1.ps

Graphical techniques for selecting variables for time series.. - Marriott, Pettitt (Correct)
Graphical techniques for selecting variables for **time series** data. By J. M. Marriott 1 The Nottingham
techniques for selecting variables for **time series** data. By J. M. Marriott 1 The Nottingham Trent
that can capture both stochastic and deterministic **trend**, seasonality and serial correlation. We propose
www.math.fsc.qut.edu.au/papers/tsx6.ps.gz

Neural Learning of Chaotic Dynamics: The Error.. - Rembrandt Bakker.. (1997) (Correct) (4 citations)
to identify chaotic dynamics from a single measured **timeseries**. The algorithm has four special features: 1.
The state of the system is extracted from the **time-series** using delays, followed by weighted Principal
www.neci.nj.nec.com/homepages/giles/papers/UMD-CS-TR-3843.neural.learning.chaotic.dynamics.ps.Z

Another Look At Swedish Business Cycles, 1861-1988 - Joakim Skalin, Timo Teräsvirta (1996) (Correct)
(2 citations)
paper considers nine long Swedish macroeconomic **time series** whose business cycle properties were
considers nine long Swedish macroeconomic **time series** whose business cycle properties were discussed by
amadeus.wiwi.hu-berlin.de/pub/papers/sfb373/sfb1996/dpsfb960096.ps.Z

Time-Series Similarity Problems and Well-Separated.. - Bollobas, Das.. (1998) (Correct) (6 citations)
**Time-Series** Similarity Problems and Well-Separated
**Time-Series** Similarity Problems and Well-Separated Geometric
similar (i.e. they react similarly to changing **market** conditions) even though one fluctuates near $30
www.almaden.ibm.com/cs/quest/papers/cg97_expanded.ps

Evaluating Neural Network Predictors by Bootstrapping - LeBaron, Weigend (1994) (Correct) (6 citations)
exhibit the method in the context of multi-variate **time series** prediction on financial data from the New
the method in the context of multi-variate **time series** prediction on financial data from the New York
held-out test set that includes the 1987 stock **market** crash. We also compare the performance of the
wueconb.wustl.edu:8089/eps/fin/papers/9411/9411002.ps.gz

Do We Often Find ARCH Because Of Neglected Outliers? - Franses, van Dijk (Correct)
of outliers is supposed to correspond with **time**-varying volatility in financial indicators, there
This phenomenon is observed in particular for **series** which are sampled daily or weekly. Since this
and weekly data on 22 exchange rates and 13 stock **market** indices using the standard Lagrange Multiplier
www.eur.nl/few/eb/papers/../pub/ei9706.ps

Learning to Classify Sensor Data - Manganaris (1995) (Correct)
Data from sensors are usually made available over **time** and are classified according to the behavior they
problem of classifying finite, univariate, **time series** that are governed by unknown deterministic
www.vuse.vanderbilt.edu/~stefanos/stefanos/ltcsd.ps

Conditional Minimum Volume Predictive Regions For Stochastic.. - Polonik, Yao (1999) (Correct)
by interval/region prediction in nonlinear **time series**, we propose a minimum volume predictor
statlab.uni-heidelberg.de/pub/reports/by.series/report.15.ps

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